

A Generative Constraint Model for Optimizing Software Deployment

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Motivation

- More and more functions in today's cars involve electronics and software
- 80-90 percent of the new innovative features are realized by distributed embedded systems
- Today's upper class cars contain up to 80 ECUs (Electronic Control Units)
- Even highly safety critical mechanical and hydraulic control systems will be replaced by electronic components

Software Deployment

- Today's embedded automotive software is highly distributed
- The automotive industry devotes increasing efforts to develop tools for automated software deployment.
- The underlying foundations comprise techniques like genetic algorithms and clustering techniques.
- Measuring the quality of the deployment process: Metrics

Metrics

Measure the quality of the deployment process

- **Bus-load:** The system's ECUs are connected by means of data-pipes or buses.
 - It is a critical metric of the system. It's value should be as low as possible. Otherwise: bottlenecks.
- **Resources:** The number and type of ECUs required to implement the functionalities.
- **Costs:** The required financial effort for a configuration.

Function Blocks

- Function block: a specific task from a software functionality of the system.
- Example: The Signaling Function
 - 1 Hit the commutator (Change State)
 - 2 Select the targeted component (Observer)
 - 3 Start blinking
- The function blocks have to exchange data \Rightarrow The network of function blocks.
- Each branch from the network has a weight, the communication frequency (CF).



Clusters

- Cluster = a set of function blocks grouped together.
- To each cluster corresponds an ECU for execution.
- Clusters should contain optimal function distribution.
- Apply quality criteria in order to create the best clustering schema.

Constraint Satisfaction Problem

- A *Constraint Satisfaction Problem* (CSP) is a tuple (V, D, CO) .

1. $\{ \text{var}_1 = (x_0 < y_0) \};$
2. $\text{min}_1 = x_0;$
3. $\text{min}_2 = y_0; \}$

Variables: $V = \{ \text{var}_1, x_0, y_0, \text{min}_1, \text{min}_2 \}$

Domains: $D = \{ D(x) = N | x \in V \}$

Constraints:

$$CO = \left\{ \begin{array}{l} \text{var}_1 = (x_0 < y_0), \\ \text{min}_1 = x_0, \text{min}_2 = y_0 \end{array} \right\}$$

- State of the art constraint solvers are available for solving CSPs.

CSP Partitioning

- The partitioning problem: How to group the function blocks into clusters?
- Transform the problem of cluster partitioning into a CSP:
 - Use the available resources
 - Use Quality Functions
 - Use the cost limitations

Resource Parametrization

- Conversion to CSP \Rightarrow defining the tuple (V, D, CO)
 - Function blocks, $F = \{f_1 \dots f_t\}$.
 - The communication frequency sets $CF = \{CF_1 \dots CF_t\}$,
 $CF_i = \{cf_{i1}, \dots, cf_{it}\}$.
 - Available set of ECUs $ECU = \{ECU_1 \dots ECU_k\}$
characterized by:
 - Memory: $MEM = \{mem_1 \dots mem_k\}$
 - Processing power: $PROC = \{proc_1 \dots proc_k\}$

Resource Constraints

- 1 The overall memory consumption of the function blocks is smaller or equal to the available memory:
- 2 Maximal function block memory constraint
- 3 Each ECU must have enough resources to accommodate the assigned cluster
- 4 A function block is deployed on a single ECU only
- 5 Any function *deploy* that distributes all functional blocks f_i on *max* ECUs is a solution.

Timing Constraints

- We assume that:
 - There is only a single path from the source x to the sink y
 - That there are no loops in the network.
- Then we must have:
 - 1 $t_{xy} \leq \sum_i t_{wc}(f_i) + t_{wc}(B_i) + t_{wc}(G_i)$.
 - 2 $SF: TSK = \{t_1, \dots, t_k\} \iff \sum_i time(t_i) \leq T_{SF}$
 - 3 t_{ECU_i} : the time that a ECU needs to execute a task t_i then
$$t_{ECU_i} \leq time(t_i).$$
- For every timing requirement $Req_{(a,b)}$ we instantiate these constraints.

Quality Constraints

- The most important factor when we partition the function blocks into clusters.
- We use predefined quality functions.
- Constraint System building: output values that no cluster is allowed to exceed.
- Each quality function receives as input parameter the *CF* set.

Quality Functions

- The External-Internal Ratio: ratio between the external and the internal costs must be as low as possible.

$$\forall C_i, i \in [1, c] \frac{E_i}{I_i}$$

- The Davies Bouldin Criteria: shows a good partitioning when the factor is as low as possible.

$$DB = \frac{1}{c} \sum_{i=1}^c \max_{j \neq i} \left[\frac{\text{diam}(C_i) + \text{diam}(C_j)}{d(C_i, C_j)} \right]$$

- The Modularization Factor: indicator of a compact clustering. Should be as high as possible.

$$MF = \frac{\sum_i I_i}{\sum_i \frac{N_i(N_i - 1)}{2}} - \frac{\sum_{i < j} E(C_i, C_j)}{\sum_{i < j} N_i N_j}$$

- The SILHOUETTE factor: correctness of the distribution of a function f_i within a cluster C_i with respect to a neighbor node C_j . A good factor is close to 1 :

$$Sh(f_i) = \frac{d(f_i, C_j) - d(f_i, C_i)}{\max(d(f_i, C_j), d(f_i, C_i))}$$

- The Cluster Load Deviation: as low as possible. In a good partitioning similar number of function blocks in all clusters:

$$CLD = \sqrt{\frac{1}{c-1} \sum_{i=1}^c (N_i - \bar{N})^2}$$

Cost Constraints

- Built based on the system's cost criteria.
- Each ECU has a price and a performance description associated to it.
- The Price Constrain: Distribute the function block set such that the total cost of the ECUs is smaller than a given price.
- The Bus Load Constraint: At any given moment the bus load should not exceed a critical imposed value.

$$CSP = RCS \cup QCS \cup CCS \cup TCS.$$

Generative Deployment of Clusters

- Different types of ECUs on which we must assign the clusters.
- The requirements of this cluster are generated by means of the generative constraints.
- Choose from the set of ECUs the cheapest ECU and try to deploy the cluster.
- If after replacing the meta variables with the concrete variables, an inconsistency happens \implies
Choose the next best ECU by using the nogoods provided by the consistency checker

Conclusions

- (CSP) representations are successfully used in diverse areas from configuration and reconfiguration of large systems.
- We outline a novel modeling approach that allows deployment of embedded automotive software.
- The Model-based approach allows for automatic deployment of software functions in a resource-constrained software system.
- Computation of a valid solution satisfying the outlined criteria by relying on standard CSP solvers.
- We do not generate all the cluster combinations but rely on the the first n , solutions that the CSP solver comes up with.

Thank You!

Q & A